# Dirichlet draws are sparse with high probability

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#### Abstract

This note provides an elementary proof of the folklore fact that draws from a Dirichlet distribution (with parameters less than 1) are typically sparse (most coordinates are small).

# 1 Bounds

Let  $Dir(\alpha)$  denote a Dirichlet distribution with all parameters equal to  $\alpha$ .

**Theorem 1.1.** Suppose  $n \geq 2$  and  $(X_1, \ldots, X_n) \sim \text{Dir}(1/n)$ . Then, for any  $c_0 \geq 1$  satisfying  $6c_0 \ln(n) + 1 < 3n$ ,

$$\Pr\left[\left|\left\{i: X_i \ge \frac{1}{n^{c_0}}\right\}\right| \le 6c_0 \ln(n)\right] \ge 1 - \frac{1}{n^{c_0}}.$$

The parameter is taken to be 1/n, which is standard in machine learning. The above theorem states that (with high probability) as the exponent on the sparsity threshold grows linearly  $(n^{-1}, n^{-2}, n^{-3}, \ldots)$ , the number of coordinates above the threshold cannot grow faster than linearly  $(6 \ln(n), 12 \ln(n), 18 \ln(n), \ldots)$ .

The above statement can be parameterized slightly more finely, exposing more tradeoffs than just the threshold and number of coordinates.

**Theorem 1.2.** Suppose  $n \ge 1$  and  $c_1, c_2, c_3 > 0$  with  $c_2 \ln(n) + 1 < 3n$ , and  $(X_1, ..., X_n) \sim \text{Dir}(c_1/n)$ ; then

$$\Pr[|\{i: X_i \ge n^{-c_3}\}| \le c_2 \ln(n)] \ge 1 - \frac{1}{e^{1/3}} \left(\frac{1}{n}\right)^{\frac{c_2}{3} - c_1 c_3} - \frac{1}{e^{4/9}} \left(\frac{1}{n}\right)^{\frac{4c_2}{9}}.$$

The natural question is whether the factor  $\ln(n)$  is an artifact of the analysis; simulation experiments with Dirichlet parameter  $\alpha = 1/n$ , summarized in Figure 1a, exhibit both the  $\ln(n)$  term, and the linear relationship between sparsity threshold and number of coordinates exceeding it.

The techniques here are loose when applied to the case  $\alpha = o(1/n)$ . In particular, Figure 1b suggests  $\alpha = 1/n^2$  leads to a single nonsmall coordinate with high probability, which is stronger than what is captured by the following theorem.

**Theorem 1.3.** Suppose  $n \geq 3$  and  $(X_1, \ldots, X_n) \sim \text{Dir}(1/n^2)$ ; then

$$\Pr[|\{i: X_i \ge n^{-2}\}| \le 5] \ge 1 - e^{2/e - 2} - e^{-8/3} \ge 0.64.$$

Moreover, for any function  $g: \mathbb{Z}_{++} \to \mathbb{R}_{++}$  and any n satisfying  $1 \le \ln(g(n)) < 3n - 1$ ,

$$\Pr[|\{i: X_i \ge n^{-2}\}| \le \ln(g(n))] \ge 1 - e^{2/e - 1/3} \left(\frac{1}{g(n)}\right)^{1/3} - e^{-4/9} \left(\frac{1}{g(n)}\right)^{4/9}.$$

(Take for instance g to be the inverse Ackermann function.)

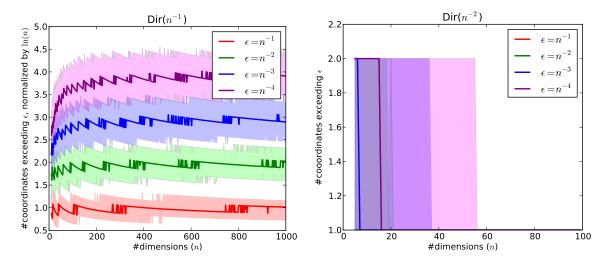


Figure 1: For each Dirichlet parameter choice  $\alpha \in \{n^{-1}, n^{-2}\}$  and each number of dimensions n (horizontal axis), 1000 Dirichlet distributions were sampled. For each trial, the number of coordinates exceeding each of 4 choices of threshold were computed. In the case of  $\alpha = n^{-1}$ , these counts were then scaled by  $\ln(n)$  to better coordinate with the suggested trends in Theorems 1.1 and 1.2. Finally, these counts values (for each  $(n, \epsilon)$ ) were converted into quantile curves (25%-75%).

# 2 Proofs

Theorems 1.1 to 1.3 are established via the following lemma.

**Lemma 2.1.** Let reals  $\epsilon \in (0,1]$  and  $\alpha > 0$  and positive integers k,n be given with k+1 < 3n. Let  $(X_i, \ldots, X_n) \sim \text{Dir}(\alpha)$ . Then

$$\Pr[|\{i: X_i \ge \epsilon\}| \le k] \ge 1 - \epsilon^{-n\alpha} e^{-(k+1)/3} - e^{-4(k+1)/9}.$$

The proof avoids dependencies between the coordinates of a Dirichlet draw via the following alternate representation. Throughout the rest of this section, let  $Gamma(\alpha)$  denote a Gamma distribution with parameter  $\alpha$ .

**Lemma 2.2.** (See for instance Balakrishnan and Nevzorov (2003, Equation 27.17).) Let  $\alpha > 0$  and  $n \ge 1$  be given. If  $(X_1, \ldots, X_n) \sim \text{Dir}(\alpha)$  and  $\{Y_i\}_{i=1}^n$  are n i.i.d. copies of Gamma( $\alpha$ ), then

$$(X_1,\ldots,X_n) \stackrel{d}{=} \left\{ \frac{Y_i}{\sum_{i=1}^n Y_i} \right\}.$$

Before turning to the proof of Lemma 2.1, one more lemma is useful, which will allow a control of the Gamma distribution's cdf.

**Lemma 2.3.** For any  $\alpha > 0$ ,  $c \geq 0$ , and  $z \geq 1$ ,

$$\Pr[\operatorname{Gamma}(\alpha) \le zc] \le z^{\alpha} \Pr[\operatorname{Gamma}(\alpha) \le c].$$

*Proof.* Since  $e^{-zx} \le e^{-x}$  for every  $x \ge 0$  and  $z \ge 1$ ,

$$\begin{split} \Pr[\mathrm{Gamma}(\alpha) \leq zc] &= \frac{1}{\Gamma(\alpha)} \int_0^{zc} e^{-x} x^{\alpha - 1} dx \\ &= \frac{1}{\Gamma(\alpha)} \int_0^c e^{-zx} (zx)^{\alpha - 1} z dx \\ &\leq \frac{z^\alpha}{\Gamma(\alpha)} \int_0^c e^{-x} x^{\alpha - 1} dx \\ &= z^\alpha \Pr[\mathrm{Gamma}(\alpha) \leq c]. \end{split}$$

Proof of Lemma 2.1. Since  $z \mapsto \Pr[\operatorname{Gamma}(\alpha) \geq z]$  is continuous and has range [0, 1], choose  $c \geq 0$  so that

$$\Pr[\operatorname{Gamma}(\alpha) > c] = \Pr[\operatorname{Gamma}(\alpha) \ge c] = \frac{k+1}{3n}, \tag{2.4}$$

where (k+1)/(3n) < 1. By this choice and Lemma 2.3,

$$\Pr[\operatorname{Gamma}(\alpha) \le c/\epsilon] \le \epsilon^{-\alpha} \Pr[\operatorname{Gamma}(\alpha) \le c] = \epsilon^{-\alpha} \left(1 - \frac{k+1}{3n}\right) \le \epsilon^{-\alpha} e^{-(k+1)/(3n)}. \tag{2.5}$$

Now let  $\{Y_i\}_{i=1}^n$  be n i.i.d. copies of  $Gamma(\alpha)$ . Define the events

$$A := [\exists i \in [n] \cdot Y_i \ge c/\epsilon]$$
 and  $B := [|\{i \in [n] : Y_i \le c\}| \ge n - k]$ .

The remainder of the proof will establish a lower bound on  $\Pr(A \wedge B)$ . To see that this finishes the proof, define  $S := \sum_i Y_i$ ; since event A implies that  $S \ge c/\epsilon$ , it follows that  $Y_i \le c$  implies  $Y_i/S \le \epsilon$ . Consequently, events A and B together imply that  $Y_i/S \le \epsilon$  for at least n-k choices of i. By Lemma 2.2, it follows that  $\Pr(A \wedge B)$  is a lower bound on the event that a draw from  $\operatorname{Dir}(\alpha)$  has at least n-k coordinates which are at most  $\epsilon$ .

Returning to task, note that

$$\Pr(A \land B) = 1 - \Pr(\neg A \lor \neg B) \ge 1 - \Pr(\neg A) - \Pr(\neg B). \tag{2.6}$$

To bound the first term, by eq. (2.5),

$$\Pr(\neg A) = \Pr[\forall i \in [n] \cdot Y_i < c/\epsilon] = \Pr[Y_1 \le c/\epsilon]^n \le \epsilon^{-n\alpha} e^{-(k+1)/3}. \tag{2.7}$$

For the second term, define indicator random variables  $Z_i := [Y_i > c]$ , whereby

$$\mathbb{E}(Z_i) = \Pr[Z_i = 1] = \Pr[Y_i > c] = \Pr[Y_i \ge c] = \frac{k+1}{3n}.$$

Then, by a multiplicative Chernoff bound (Kearns and Vazirani, 1994, Theorem 9.2),

$$\Pr(\neg B) = \Pr[|\{i \in [n] : Y_i > c\}| \ge k + 1] = \Pr\left[\sum_i Z_i \ge 3n\mathbb{E}(Z_i)\right] \le \exp(-4n\mathbb{E}(Z_i)/3).$$
 (2.8)

Inserting (2.7) and (2.8) into the lower bound on  $Pr(A \wedge B)$  in (2.6),

$$\Pr(A \wedge B) \ge 1 - e^{-n\alpha} e^{-(k+1)/3} - e^{-4(k+1)/9}.$$

Proof of Theorem 1.2. Instantiate Lemma 2.1 with  $k = c_2 \ln(n)$ ,  $\alpha = c_1/n$ , and  $\epsilon = n^{-c_3}$ .

Proof of Theorem 1.1. Instantiate Theorem 1.2 with  $c_1 = 1$ ,  $c_2 = 6c_0$ ,  $c_3 = c_0$ , and note

$$\frac{1}{e^{1/3}} \left(\frac{1}{n}\right)^{c_0} + \frac{1}{e^{4/9}} \left(\frac{1}{n}\right)^{\frac{8c_0}{3}} \leq \frac{1}{n^{c_0}} \left(\frac{1}{e^{1/3}} + \frac{1}{e^{4/9}} \left(\frac{1}{2}\right)^{\frac{5c_0}{3}}\right) \leq \frac{1}{n^{c_0}}.$$

Proof of Theorem 1.3. Define the function  $f(z) := z^{-z}$  over  $(0, \infty)$ . Note that  $f'(z) = -(\ln(z) + 1)z^{-z}$ , which is positive for z < 1/e, zero at z = 1/e, and negative thereafter; consequently,  $\sup_{z \in (0,\infty)} f(z) = f(1/e) = e^{1/e}$ . As such, instantiating Lemma 2.1 with  $\epsilon = n^{-2}$ ,  $\alpha = n^{-3}$ , and any k < 3n - 1 gives

$$\Pr[|\{i: X_i \ge n^{-2}\}| \le k] \ge 1 - n^{2/n} e^{-(k+1)/3} - e^{-4(k+1)/9}$$

$$> 1 - e^{2/e} e^{-(k+1)/3} - e^{-4(k+1)/9}.$$

Plugging in  $k \in \{5, \ln(g(n))\}$  gives the two bounds.

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### References

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